

Diatom.assembly

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Set working space

Load libraries

```
rm(list=ls())
library(vegan)
library(ggplot2)
library(dplyr)
library(readxl)
library(xlsx)
library(cowplot)
library(ape)
library(phyloseq)
library(stringr)
library(iCAMP)
library(NST)
library(dplyr)
library(tidyr)
library(grid)
library(tidyverse)
library(ggvenn)
library("ggVennDiagram")
library(pheatmap)
library(cowplot)
library("pgirmess")
library(ggsignif)
```

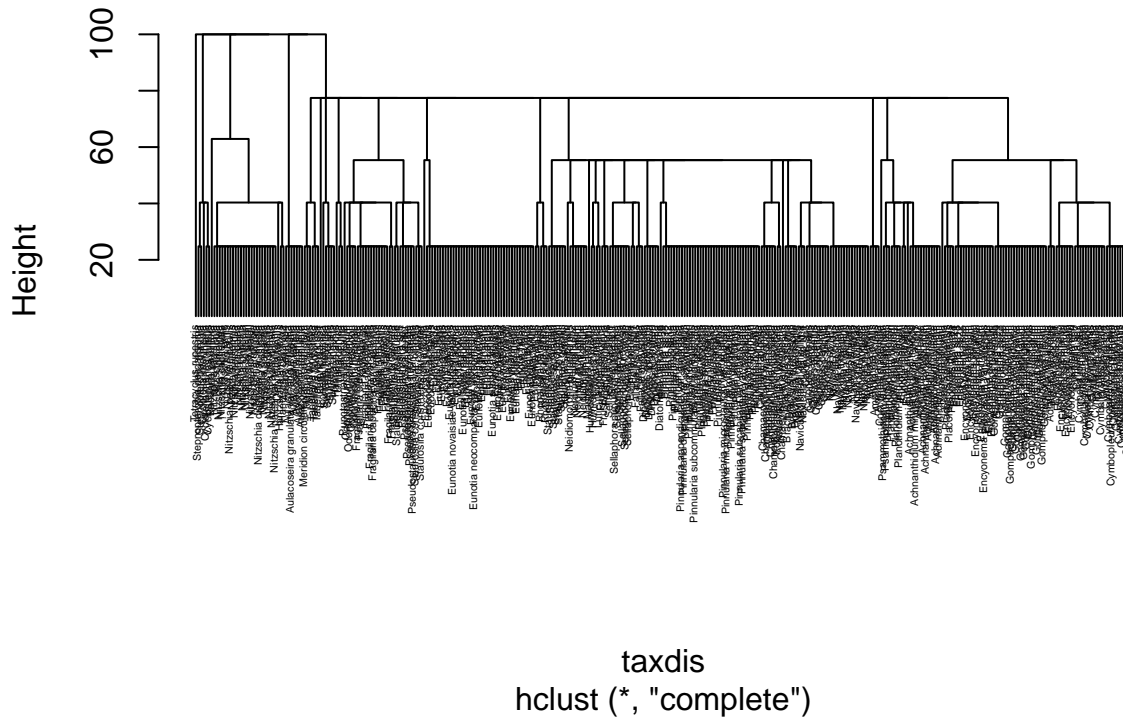
Community assembly processes

Load taxonomic tree dataset and deterministic processes (bNTI)

```
taxon_name=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="Taxonomic tree",na=""))
rownames(taxon_name)=taxon_name$Species
Abun_rel=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="Relative abundance",na=""))
comm<-t(sqrt(Abun_rel[8:360]))
TaxonID<-row.names(comm)

# Taxonomic distances from a classification table with variable step lengths.
taxdis <- taxa2dist(taxon_name, varstep=TRUE)
plot(hclust(taxdis), hang = -1, cex=0.3)
```

Cluster Dendrogram



```
dist.df=data.frame(as.matrix(taxdis))
```

```
# bNTI calculation
#row.names(dist.df)<-names(dist.df)
#set.seed(1)
#NTI.out = bNTIn.p(t(comm), dist.df, nworker = 4, memo.size.GB = 16, weighted = TRUE,
#                   exclude.consp = FALSE, rand = 1000, output.bMNTD = TRUE,
#                   sig.index = "bNTI",
#                   unit.sum = NULL, correct.special = FALSE, detail.null = FALSE,
#                   special.method = "MNTD")

#save(file="bNTI_bioocult_20241202.RData",NTI.out)
```

Load relative abundance dataset and Raup crick (RC) calculation

```
load("data\\bNTI_biocult_20241202.RData")
metadata=data.frame(read_xlsx("data\\Table S1.xlsx",sheet="Metadata",na=""))
Abun_rel=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="Relative abundance",na=""))

rownames(Abun_rel)=Abun_rel$Station
rownames(metadata)=metadata$Station

xy <- t(combn(colnames(NTI.out$index), 2))
tmp = data.frame(xy, BNTI = NTI.out$index[xy])
tmp_merge=merge(tmp,metadata[,1:4],by.x="X1",by.y="Station")
indice=merge(tmp_merge,metadata[,1:4],by.x="X2",by.y="Station")
```

```

indice=indice[indice$Local.x==indice$Local.y,]
indice$proc="Stochastic"
indice$proc[indice$BNTI< (-2)]= "Homogeneous selection"
indice$proc[indice$BNTI>2]= "Heterogeneous selection"
indice$index=paste0(indice$X2,".",indice$X1)

```

Mire inside analysis (Local scale)

RC mire inside

```

comm<-t(sqrt(Abun_rel[8:360]))
TaxonID<-row.names(comm)
meta.groupi<-data.frame(cluster=Abun_rel[,c("Local")])
row.names(meta.groupi)<-Abun_rel$Station

### taxonomic normalized stochasticity ratio
set.seed(1)
tnst=tNST(comm=t(comm),meta.com=NULL,group=meta.groupi,
           dist.method="bray", abundance.weighted=TRUE, rand=1000,
           output.rand=TRUE, nworker=8, LB=FALSE, null.model="PF",
           between.group=TRUE, SES=TRUE, RC=TRUE)

#to extract only the comparisons within each mire
nst.temp<-tnst[["index.pair.grp"]]
nst.temp$index=paste0(nst.temp$name1,".",nst.temp$name2)

#RC Percentage
nst.temp$category<- "bRC<|0.95|"
nst.temp$category[nst.temp$RC.ij.bray>0.95]<- "bRC>0.95"
nst.temp$category[nst.temp$RC.ij.bray< (-0.95)]<- "bRC<-0.95"

#To extract stochastic data
stoch_cate=nst.temp[,c(2,3,10,11,12)]
stoch_cate$proc<- "Ecological drift"
stoch_cate$proc[stoch_cate$RC.ij.bray>0.95]<- "Dispersal limitation"
stoch_cate$proc[stoch_cate$RC.ij.bray< (-0.95)]<- "Homogenizing dispersal"

procesos=merge(indice,stoch_cate,by="index")

procesos$process<-NA
procesos$process=procesos$proc.x
procesos$process=ifelse(procesos$process=="Stochastic",procesos$proc.y,procesos$proc.x)

agg.bRC<-aggregate(RC.ij.bray~process+Local.x,data=procesos,length)
agg.bRC<-agg.bRC%>% complete(process,Local.x)#complete when there is nothing
agg.bRC$RC.ij.bray[is.na(agg.bRC$RC.ij.bray)]<-0#if NA replace with 0

agg.bRC$category<-as.factor(agg.bRC$process) #Set it as factor

#Percentages
gg.tmp<-aggregate(RC.ij.bray~Local.x,data=procesos,length)
gg.bRC_percentage<-merge(agg.bRC,gg.tmp,by="Local.x")

```

```
gg.bRC_percentage$perc.RC<-gg.bRC_percentage$RC.ij.bray.x/gg.bRC_percentage$RC.ij.bray.y
```

Display panel Figure S3 A

```
cbp1 <- c( "#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
           "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

#define factors and added national parks in each mire
agg.bRC_all=merge(agg.bRC,unique(procesos[,6:7]),by = "Local.x")

agg.bRC_all$group<-factor(agg.bRC_all$Local.x,
                          c("PCCO", "PCER", "PLIO", "PRON", "PSVB",
                            "OBDC", "OBUS", "OOTA", "OOTS",
                            "GBAD", "GMHO", "GNAJ", "GRAS",
                            "NCLV", "NCPG", "NDAL", "NDCM", "NTCC"))
agg.bRC_all$proc<-factor(agg.bRC_all$process,
                         c( "Ecological drift", "Dispersal limitation",
                           "Homogenizing dispersal",
                           "Heterogeneous selection", "Homogeneous selection"))

agg.bRC_all$Landscape.x<-factor(agg.bRC_all$Landscape.x,
                                c("Picos de Europa", "Ordesa y Monte Perdido",
                                  "Guadarrama", "Sierra Nevada"))

prop.bRC<-agg.bRC_all%>%
  ggplot(aes(x=group, y = RC.ij.bray, fill = proc)) +
  geom_col(position = "fill")+
  scale_fill_manual(values = c(cb1[1:6]),name="")+
  labs(title="",x=NULL,y="Community assembly processes (%)")+
  theme_bw()+
  theme(panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.y = element_text(size=10, color="black",hjust=1,angle = 0),
        axis.text.x = element_blank(),axis.title.x = element_blank(),
        text = element_text(family="serif"))+
  facet_wrap(~Landscape.x,scales = "free_x",nrow = 1)+
  theme(plot.margin = unit(c(0.1,0.15, 0.1,0), "cm"))+
  theme(legend.key.size = unit(0.7, "lines"),
        legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),
        legend.position = "right",legend.box = "horizontal",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.2,"cm"),legend.text=element_text(size=8),
        legend.title=element_text(size=6))+
  theme(panel.spacing = unit(0.1, "cm"),strip.text.x = element_blank(),
        strip.text.y = element_text(size = 6, color = "black"),
        panel.border = element_rect(color = "grey", fill = NA))+
  guides(fill=guide_legend(ncol = 2))
legenda_prop.bRC=get_plot_component(prop.bRC, 'guide-box-right', return_all = TRUE)
prop.bRC=prop.bRC+theme(legend.position = "none")
```

National parks inside analysis (Landscape scale)

Display panel Figure S3 B

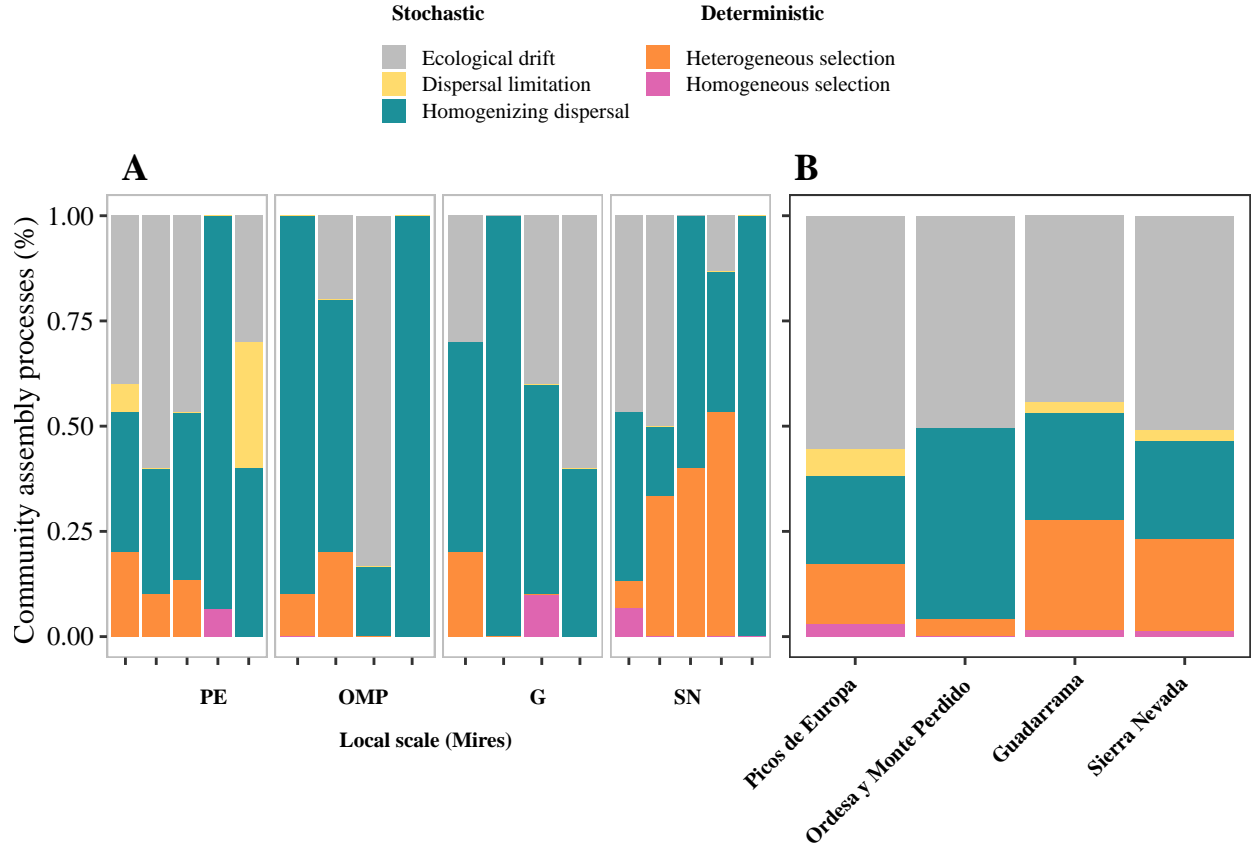
```
cbp1 <- c("#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

agg.bRC_np$group<-factor(agg.bRC_np$Landscape.x,
                        c("Picos de Europa", "Ordesa y Monte Perdido",
                          "Guadarrama", "Sierra Nevada"))
agg.bRC_np$proc<-factor(agg.bRC_np$process,
                      c("Ecological drift", "Dispersal limitation",
                        "Homogenizing dispersal",
                        "Heterogeneous selection", "Homogeneous selection"))

prop.bRC_np<-agg.bRC_np%>%
  ggplot(aes(x = group, y = RC.ij.bray, fill = proc)) +
  geom_col(position = "fill")+scale_fill_manual(values = c(cbp1[1:5]),
                                                name="Community assembly process")+
  labs(title="", x=NULL, y=expression(beta*"RC (%) "))+
  theme_bw()+
  theme(panel.grid.minor = element_blank(), panel.grid.major = element_blank(),
        axis.text.x = element_blank(), axis.title.x = element_blank(),
        axis.title.y = element_blank(),
        axis.text.y = element_blank(), text = element_text(family="serif"))+
  theme(plot.margin = unit(c(0.1, 0.15, 0.1, 0), "cm"))+
  theme(legend.key.size = unit(1, "lines"),
        legend.margin=margin(0, 1, 0, 1), legend.box.margin=margin(-10, -10, -10, -10),
        legend.position = "none", legend.justification="bottom",
        legend.box = "horizontal",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.3, "cm"), legend.text=element_text(size=6))
```

Display Figure S2

Figure S2: A different approach, grouping multiple landscape estimates, shows the variability in the assembly processes of diatom communities at the regional scale. Representation of the various combinations of national parks (A, B, C). Percentage of community assembly processes in the six different combinations (D, E, F). The code represented the national parks' names: Ordesa y Monte Perdido (OMP), Guadarrama (G), Picos de Europa (PE), and Sierra Nevada (SN).



Comparison between pairs national parks

```
comm_p<-t(sqrt(Abun_rel[8:360]))
TaxonID_p<-row.names(comm_p)
meta.groupi_p<-data.frame(cluster=Abun_rel[,c("Landscape")])
row.names(meta.groupi_p)<-Abun_rel$Station

#taxonomic normalized stochasticity ratio
set.seed(1)
tnst_p=tNST(comm=t(comm_p),meta.com=NULL,group=meta.groupi_p,
             dist.method="bray", abundance.weighted=TRUE, rand=1000,
             output.rand=TRUE, nworker=8, LB=FALSE, null.model="PF",
             between.group=TRUE, SES=TRUE, RC=TRUE)

#to extract only the comparisons between each national park
nst.temp_p<-tnst_p[["index.pair"]]
nst.temp_p$index=paste0(nst.temp_p$name1,".",nst.temp_p$name2)

#RC percentage
nst.temp_p$category<-"bRC<|0.95|"
nst.temp_p$category[nst.temp_p$RC.bray>0.95]<-"bRC>0.95"
nst.temp_p$category[nst.temp_p$RC.bray<(-0.95)]<-"bRC<-0.95"

#To extract stochastic data
stoch_cate_p=nst.temp_p[,c(1,2,12,13,14)]
```

```

stoch_cate_p$proc<-"Ecological drift"
stoch_cate_p$proc[stoch_cate_p$RC.bray>0.95]<-"Dispersal limitation"
stoch_cate_p$proc[stoch_cate_p$RC.bray< (-0.95)]<-"Homogenizing dispersal"

xy_p <- t(combn(colnames(NTI.out$index), 2))
tmp_p = data.frame(xy_p, BNTI = NTI.out$index[xy_p])
tmp_merge_p=merge(tmp_p,metadata[,1:4],by.x="X1",by.y="Station")
indice_p=merge(tmp_merge_p,metadata[,1:4],by.x="X2",by.y="Station")
indice_p=indice_p[indice_p$Landscape.x!=indice_p$Landscape.y,]
indice_p$proc="Stochastic"
indice_p$proc[indice_p$BNTI< (-2)]= "Homogeneous selection"
indice_p$proc[indice_p$BNTI>2]= "Heterogeneous selection"
indice_p$index=paste0(indice_p$X2,".",indice_p$X1)

procesos_p=merge(indice_p,stoch_cate_p,by="index")

procesos_p$process<-NA
procesos_p$process=procesos_p$proc.x
procesos_p$process=ifelse(procesos_p$process=="Stochastic",
                           procesos_p$proc.y,procesos_p$proc.x)
procesos_p$index.np=paste0(procesos_p$Landscape.x,".",procesos_p$Landscape.y)
unique(procesos_p$index.np)

## [1] "Sierra Nevada.Guadarrama"
## [2] "Ordesa y Monte Perdido.Guadarrama"
## [3] "Picos de Europa.Guadarrama"
## [4] "Picos de Europa.Sierra Nevada"
## [5] "Sierra Nevada.Ordesa y Monte Perdido"
## [6] "Picos de Europa.Ordesa y Monte Perdido"

agg.bRC_p<-aggregate(RC.bray~process+index.np,data=procesos_p,length)
agg.bRC_p$category<-as.factor(agg.bRC_p$process) #Set it as factor

#percentage
gg.tmp_p<-aggregate(RC.bray~index.np,data=procesos_p,length)
gg.bRC_percentage_p<-merge(agg.bRC_p,gg.tmp_p,by="index.np")

gg.bRC_percentage_p$perc.RC<-
  gg.bRC_percentage_p$RC.bray.x/gg.bRC_percentage_p$RC.bray.y

```

Display panel Figure 3B

```

cbp1 <- c("#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

agg.bRC_p$group<-factor(agg.bRC_p$index.np,
                        c("Picos de Europa.Ordesa y Monte Perdido",
                          "Picos de Europa.Guadarrama",
                          "Picos de Europa.Sierra Nevada",
                          "Ordesa y Monte Perdido.Guadarrama",
                          "Sierra Nevada.Ordesa y Monte Perdido",
                          "Sierra Nevada.Guadarrama"))
agg.bRC_p$proc<-factor(agg.bRC_p$process,
                      c("Ecological drift", "Dispersal limitation",

```

```

        "Homogenizing dispersal",
        "Heterogeneous selection", "Homogeneous selection"))

RC_pe_omp=agg.bRC_p[agg.bRC_p$group=="Picos de Europa.Ordesa y Monte Perdido",]%>%
  ggplot(aes(x="", y = RC.bray, fill = proc)) +
  geom_bar(stat="identity", width=1) +
  coord_polar("y", start=0)+
  scale_fill_manual(values = c(cbp1[1:5]),name="")+
  theme_minimal()+ theme_void()+
  theme(legend.position = "none")+theme(plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))

RC_pe_g=agg.bRC_p[agg.bRC_p$group=="Picos de Europa.Guadarrama",]%>%
  ggplot(aes(x="", y = RC.bray, fill = proc)) +
  geom_bar(stat="identity", width=1) +
  coord_polar("y", start=0)+
  scale_fill_manual(values = c(cbp1[1:5]),name="")+
  theme_minimal()+ theme_void()+
  theme(legend.position = "none")+theme(plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))

RC_pe_sn=agg.bRC_p[agg.bRC_p$group=="Picos de Europa.Sierra Nevada",]%>%
  ggplot(aes(x="", y = RC.bray, fill = proc)) +
  geom_bar(stat="identity", width=1) +
  coord_polar("y", start=0)+
  scale_fill_manual(values = c(cbp1[1:5]),name="")+
  theme_minimal()+ theme_void()+
  theme(legend.position = "none")+theme(plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))

RC_omp_g=agg.bRC_p[agg.bRC_p$group=="Ordesa y Monte Perdido.Guadarrama",]%>%
  ggplot(aes(x="", y = RC.bray, fill = proc)) +
  geom_bar(stat="identity", width=1) +
  coord_polar("y", start=0)+
  scale_fill_manual(values = c(cbp1[1:5]),name="")+
  theme_minimal()+ theme_void()+
  theme(legend.position = "none")+theme(plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))

RC_sn_omp=agg.bRC_p[agg.bRC_p$group=="Sierra Nevada.Ordesa y Monte Perdido",]%>%
  ggplot(aes(x="", y = RC.bray, fill = proc)) +
  geom_bar(stat="identity", width=1) +
  coord_polar("y", start=0)+
  scale_fill_manual(values = c(cbp1[1:5]),name="")+
  theme_minimal()+ theme_void()+
  theme(legend.position = "none")+theme(plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))

```



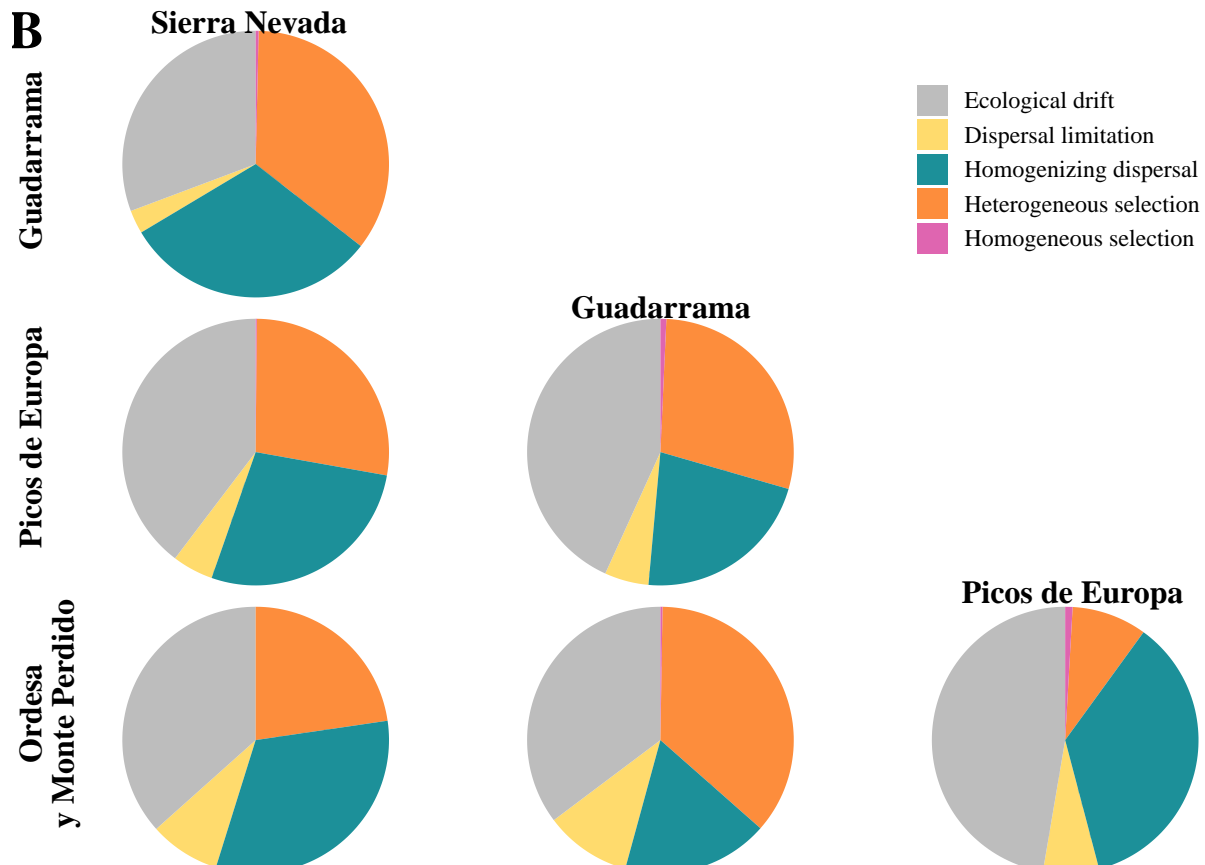
```

RC_sn_g=agg.bRC_p[agg.bRC_p$group=="Sierra Nevada.Guadarrama",]%>%
ggplot(aes(x="", y = RC.bray, fill = proc)) +
geom_bar(stat="identity", width=1) +
coord_polar("y", start=0)+
scale_fill_manual(values = c(cbp1[1:5]),name="")+
theme_minimal()+ theme_void()+
theme(text = element_text(family="serif"))+
theme(legend.position = "right",legend.key.size = unit(0.9, "lines"),
      legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),
      legend.box = "horizontal",legend.key = element_rect(color = NA, fill = NA),
      legend.spacing.x = unit(0.1,"cm"),legend.text=element_text(size=9),
      plot.margin = unit(c(-0.1,-.3,-.5,-.1),"cm"))
legenda_sn_g=get_legend(RC_sn_g)
RC_sn_g=RC_sn_g+theme(legend.position = "none")

```

Figure 3

Figure 3: Pairwise comparison in assembly processes between all landscapes (national parks). A) Description of the pairwise comparison conducted between national parks. B) The percentage of assembly processes between national parks.



Six different combination of national parks to calculate regional community assembly

First combination

```
comm_pp<-t(sqrt(Abun_rel[8:360]))
TaxonID_pp<-row.names(comm_pp)
meta.groupi_pp<-data.frame(cluster=Abun_rel[,c("Regional_1")])
row.names(meta.groupi_pp)<-Abun_rel$Station

#taxonomic normalized stochasticity ratio
set.seed(1)
tnst_pp=tNST(comm=t(comm_pp),meta.com=NULL,group=meta.groupi_pp,
             dist.method="bray", abundance.weighted=TRUE, rand=1000,
             output.rand=TRUE, nworker=8, LB=FALSE, null.model="PF",
             between.group=TRUE, SES=TRUE, RC=TRUE)

#to extract each combination of national parks
nst.temp_pp<-tnst_pp[["index.pair.grp"]]
nst.temp_pp$index=paste0(nst.temp_pp$name1,".",nst.temp_pp$name2)

#RC percentage
nst.temp_pp$category<-"bRC<|0.95|"
nst.temp_pp$category[nst.temp_pp$RC.ij.bray>0.95]<-"bRC>0.95"
nst.temp_pp$category[nst.temp_pp$RC.ij.bray< (-0.95)]<-"bRC<-0.95"

#To extract stochastic data
stoch_cate_pp=nst.temp_pp[,c(2,3,10,11,12)]
stoch_cate_pp$proc<-"Ecological drift"
stoch_cate_pp$proc[stoch_cate_pp$RC.ij.bray>0.95]<-"Dispersal limitation"
stoch_cate_pp$proc[stoch_cate_pp$RC.ij.bray< (-0.95)]<-"Homogenizing dispersal"

xy_pp <- t(combn(colnames(NTI.out$index), 2))
tmp_pp = data.frame(xy_pp, BNTI = NTI.out$index[xy_pp])
tmp_merge_pp=merge(tmp_pp,Abun_rel[,4:7],by.x="X1",by.y="Station")
indice_pp=merge(tmp_merge_pp,Abun_rel[,4:7],by.x="X2",by.y="Station")
indice_pp=indice_pp[indice_pp$Regional_1.x==indice_pp$Regional_1.y,]
indice_pp$proc="Stochastic"
indice_pp$proc[indice_pp$BNTI< (-2)]= "Homogeneous selection"
indice_pp$proc[indice_pp$BNTI>2]= "Heterogeneous selection"
indice_pp$index=paste0(indice_pp$X2,".",indice_pp$X1)

procesos_pp=merge(indice_pp,stoch_cate_pp,by="index")

procesos_pp$process<-NA
procesos_pp$process=procesos_pp$proc.x
procesos_pp$process=ifelse(procesos_pp$process=="Stochastic",
                           procesos_pp$proc.y,procesos_pp$proc.x)
procesos_pp$index.np=paste0(procesos_pp$Regional_1.x,".",
                           procesos_pp$Regional_1.y)
unique(procesos_pp$index.np)

## [1] "OMP_G.OMP_G" "PE_SN.PE_SN"
```

```
agg.bRC_pp<-aggregate(RC.ij.bray~process+index.np,data=procesos_pp,length)
agg.bRC_pp$category<-as.factor(agg.bRC_pp$process) #Set it as factor

#percentage
gg.tmp_pp<-aggregate(RC.ij.bray~index.np,data=procesos_pp,length)
gg.bRC_percentage_pp<-merge(agg.bRC_pp,gg.tmp_pp,by="index.np")

gg.bRC_percentage_pp$perc.RC<-
  gg.bRC_percentage_pp$RC.ij.bray.x/gg.bRC_percentage_pp$RC.ij.bray.y
```

Display panel Figure S2 D

```
cbp1 <- c("#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

agg.bRC_pp$group<-factor(agg.bRC_pp$index.np,
                        c("OMP_G.OMP_G", "PE_SN.PE_SN"))
agg.bRC_pp$proc<-factor(agg.bRC_pp$process,
                        c("Ecological drift", "Dispersal limitation",
                          "Homogenizing dispersal",
                          "Heterogeneous selection", "Homogeneous selection"))

prop.bRC_pp<-agg.bRC_pp%>%
  ggplot(aes(x = group, y = RC.ij.bray, fill = proc)) +
  geom_col(position = "fill")+scale_fill_manual(values = c(cbp1[1:5]),name="")+
  labs(title="",x=NULL,y="Community assembly processes (%)")+
  theme_bw()+
  theme(panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.y = element_text(size=10, color="black",hjust=1,angle = 0),
        axis.text.x = element_blank(),axis.title.x = element_blank())+
  theme(text = element_text(family = "serif"))+
  theme(plot.margin = unit(c(0.1,0.15, 0.1,0), "cm"))+
  theme(legend.key.size = unit(1, "lines"),
        legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),
        legend.position = "none",legend.box = "horizontal",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.3,"cm"),legend.text=element_text(size=6))
```

Second combination

```
comm_pp2<-t(sqrt(Abun_rel[8:360]))
TaxonID_pp2<-row.names(comm_pp2)
meta.groupi_pp2<-data.frame(cluster=Abun_rel[,c("Regional_2")])
row.names(meta.groupi_pp2)<-Abun_rel$Station

#taxonomic normalized stochasticity ratio
set.seed(1)
tnst_pp2=tNST(comm=t(comm_pp2),meta.com=NULL,group=meta.groupi_pp2,
              dist.method="bray", abundance.weighted=TRUE, rand=1000,
              output.rand=TRUE, nworker=8, LB=FALSE, null.model="PF",
              between.group=TRUE, SES=TRUE, RC=TRUE)
```

```

#to extract each combination of national parks
nst.temp_pp2<-tnst_pp2[["index.pair.grp"]]
nst.temp_pp2$index=paste0(nst.temp_pp2$name1,".",nst.temp_pp2$name2)

#RC percentage
nst.temp_pp2$category<-"bRC<|0.95|"
nst.temp_pp2$category[nst.temp_pp2$RC.ij.bray>0.95]<-"bRC>0.95"
nst.temp_pp2$category[nst.temp_pp2$RC.ij.bray< (-0.95)]<-"bRC<-0.95"

#To extract stochastic data
stoch_cate_pp2=nst.temp_pp2[,c(1,2,3,10,11,12)]
stoch_cate_pp2$proc<-"Ecological drift"
stoch_cate_pp2$proc[stoch_cate_pp2$RC.ij.bray>0.95]<-"Dispersal limitation"
stoch_cate_pp2$proc[stoch_cate_pp2$RC.ij.bray< (-0.95)]<-"Homogenizing dispersal"

xy_pp2 <- t(combn(colnames(NTI.out$index), 2))
tmp_pp2 = data.frame(xy_pp2, BNTI = NTI.out$index[xy_pp2])
tmp_merge_pp2=merge(tmp_pp2,Abun_rel[,3:7],by.x="X1",by.y="Station")
indice_pp2=merge(tmp_merge_pp2,Abun_rel[,3:7],by.x="X2",by.y="Station")
indice_pp2=indice_pp2[indice_pp2$Regional_2.x==indice_pp2$Regional_2.y,]
indice_pp2$proc="Stochastic"
indice_pp2$proc[indice_pp2$BNTI< (-2)]= "Homogeneous selection"
indice_pp2$proc[indice_pp2$BNTI>2]= "Heterogeneous selection"
indice_pp2$index=paste0(indice_pp2$X2,".",indice_pp2$X1)

procesos_pp2=merge(indice_pp2,stoch_cate_pp2,by="index")

procesos_pp2$process<-NA
procesos_pp2$process=procesos_pp2$proc.x
procesos_pp2$process=ifelse(procesos_pp2$process=="Stochastic",
                             procesos_pp2$proc.y,procesos_pp2$proc.x)
procesos_pp2$index.np=paste0(procesos_pp2$Regional_2.x,".",
                             procesos_pp2$Regional_2.y)
unique(procesos_pp2$index.np)

## [1] "SN_G.SN_G"      "PE_OMP.PE_OMP"

agg.bRC_pp2<-aggregate(RC.ij.bray~process+index.np,data=procesos_pp2,length)
agg.bRC_pp2$category<-as.factor(agg.bRC_pp2$process) #Set it as factor

#percentage
gg.tmp_pp2<-aggregate(RC.ij.bray~index.np,data=procesos_pp2,length)
gg.bRC_percentage_pp2<-merge(agg.bRC_pp2,gg.tmp_pp2,by="index.np")

gg.bRC_percentage_pp2$perc.RC<-
  gg.bRC_percentage_pp2$RC.ij.bray.x/gg.bRC_percentage_pp2$RC.ij.bray.y

```

Display panel Figure S2 E

```

cbp1 <- c("#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

agg.bRC_pp2$group<-factor(agg.bRC_pp2$index.np,
                          c("SN_G.SN_G", "PE_OMP.PE_OMP"))

```

```

agg.bRC_pp2$proc<-factor(agg.bRC_pp2$process,
                        c("Ecological drift","Dispersal limitation",
                          "Homogenizing dispersal",
                          "Heterogeneous selection","Homogeneous selection"))

prop.bRC_pp2<-agg.bRC_pp2%>%
  ggplot(aes(x = group, y = RC.ij.bray, fill = proc)) +
  geom_col(position = "fill")+scale_fill_manual(values = c(cbp1[1:5]),name="")+
  labs(title="",x=NULL,y="Percentage (%)")+
  theme_bw()+
  theme(panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.y = element_text(size=10, color="black",hjust=1,angle = 0),
        axis.text.x = element_blank(),axis.title.x = element_blank(),
        axis.title.y = element_blank())+
  theme(text = element_text(family = "serif"))+
  theme(plot.margin = unit(c(0.1,0.15, 0.1,0), "cm"))+
  theme(legend.key.size = unit(1, "lines"),
        legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),
        legend.position = "none",legend.box = "horizontal",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.3,"cm"),legend.text=element_text(size=6))

```

Third combination

```

comm_pp3<-t(sqrt(Abun_rel[8:360]))
TaxonID_pp3<-row.names(comm_pp3)
meta.groupi_pp3<-data.frame(cluster=Abun_rel[,c("Regional_3")])
row.names(meta.groupi_pp3)<-Abun_rel$Station

#taxonomic normalized stochasticity ratio
set.seed(1)
tnst_pp3=tnst(comm=t(comm_pp3),meta.com=NULL,group=meta.groupi_pp3,
              dist.method="bray", abundance.weighted=TRUE, rand=1000,
              output.rand=TRUE, nworker=8, LB=FALSE, null.model="PF",
              between.group=TRUE, SES=TRUE, RC=TRUE)

nst.temp_pp3<-tnst_pp3[["index.pair.grp"]]#extraemos dentro de los grupo
nst.temp_pp3$index=paste0(nst.temp_pp3$name1,".",nst.temp_pp3$name2)

#RC percentage
nst.temp_pp3$category<-"bRC<|0.95|"
nst.temp_pp3$category[nst.temp_pp3$RC.ij.bray>0.95]<-"bRC>0.95"
nst.temp_pp3$category[nst.temp_pp3$RC.ij.bray< (-0.95)]<-"bRC<-0.95"

#To extract stochastic data
stoch_cate_pp3=nst.temp_pp3[,c(1,2,3,10,11,12)]
stoch_cate_pp3$proc<-"Ecological drift"
stoch_cate_pp3$proc[stoch_cate_pp3$RC.ij.bray>0.95]<-"Dispersal limitation"
stoch_cate_pp3$proc[stoch_cate_pp3$RC.ij.bray< (-0.95)]<-"Homogenizing dispersal"

xy_pp3 <- t(combn(colnames(NTI.out$index), 2))
tmp_pp3 = data.frame(xy_pp3, BNTI = NTI.out$index[xy_pp3])
tmp_merge_pp3=merge(tmp_pp3,Abun_rel[,2:7],by.x="X1",by.y="Station")

```

```

indice_pp3=merge(tmp_merge_pp3,Abun_rel[,2:7],by.x="X2",by.y="Station")
indice_pp3=indice_pp3[indice_pp3$Regional_3.x==indice_pp3$Regional_3.y,]
indice_pp3$proc="Stochastic"
indice_pp3$proc[indice_pp3$BNTI< (-2)]= "Homogeneous selection"
indice_pp3$proc[indice_pp3$BNTI>2]= "Heterogeneous selection"
indice_pp3$index=paste0(indice_pp3$X2,".",indice_pp3$X1)

procesos_pp3=merge(indice_pp3,stoch_cate_pp3,by="index")

procesos_pp3$process<-NA
procesos_pp3$process=procesos_pp3$proc.x
procesos_pp3$process=ifelse(procesos_pp3$process=="Stochastic",
                             procesos_pp3$proc.y,procesos_pp3$proc.x)
procesos_pp3$index.np=paste0(procesos_pp3$Regional_3.x,".",
                             procesos_pp3$Regional_3.y)
unique(procesos_pp3$index.np)

## [1] "PE_G.PE_G"      "SN_OMP.SN_OMP"

agg.bRC_pp3<-aggregate(RC.ij.bray~process+index.np,data=procesos_pp3,length)
agg.bRC_pp3$category<-as.factor(agg.bRC_pp3$process) #Set it as factor

#percentage
gg.tmp_pp3<-aggregate(RC.ij.bray~index.np,data=procesos_pp3,length)
gg.bRC_percentage_pp3<-merge(agg.bRC_pp3,gg.tmp_pp3,by="index.np")

gg.bRC_percentage_pp3$perc.RC<-
  gg.bRC_percentage_pp3$RC.ij.bray.x/gg.bRC_percentage_pp3$RC.ij.bray.y

```

Display panel Figure S2 F

```

cbp1 <- c("#bdbdbd", "#FFDB6D", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

agg.bRC_pp3$group<-factor(agg.bRC_pp3$index.np,
                          c("PE_G.PE_G", "SN_OMP.SN_OMP"))
agg.bRC_pp3$proc<-factor(agg.bRC_pp3$process,
                         c("Ecological drift", "Dispersal limitation",
                           "Homogenizing dispersal",
                           "Heterogeneous selection", "Homogeneous selection"))

prop.bRC_pp3<-agg.bRC_pp3%>%
  ggplot(aes(x = group, y = RC.ij.bray, fill = proc)) +
  geom_col(position = "fill")+scale_fill_manual(values = c(cbp1[1:5]),name="")+
  labs(title="",x=NULL,y="Percentage (%)")+
  theme_bw()+
  theme(panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.y = element_text(size=10, color="black",hjust=1,angle = 0),
        axis.text.x = element_blank(),axis.title.x = element_blank(),
        axis.title.y = element_blank())+
  theme(text = element_text(family = "serif"))+
  theme(plot.margin = unit(c(0.1,0.15, 0.1,0), "cm"))+
  theme(legend.key.size = unit(1, "lines"),
        legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),

```

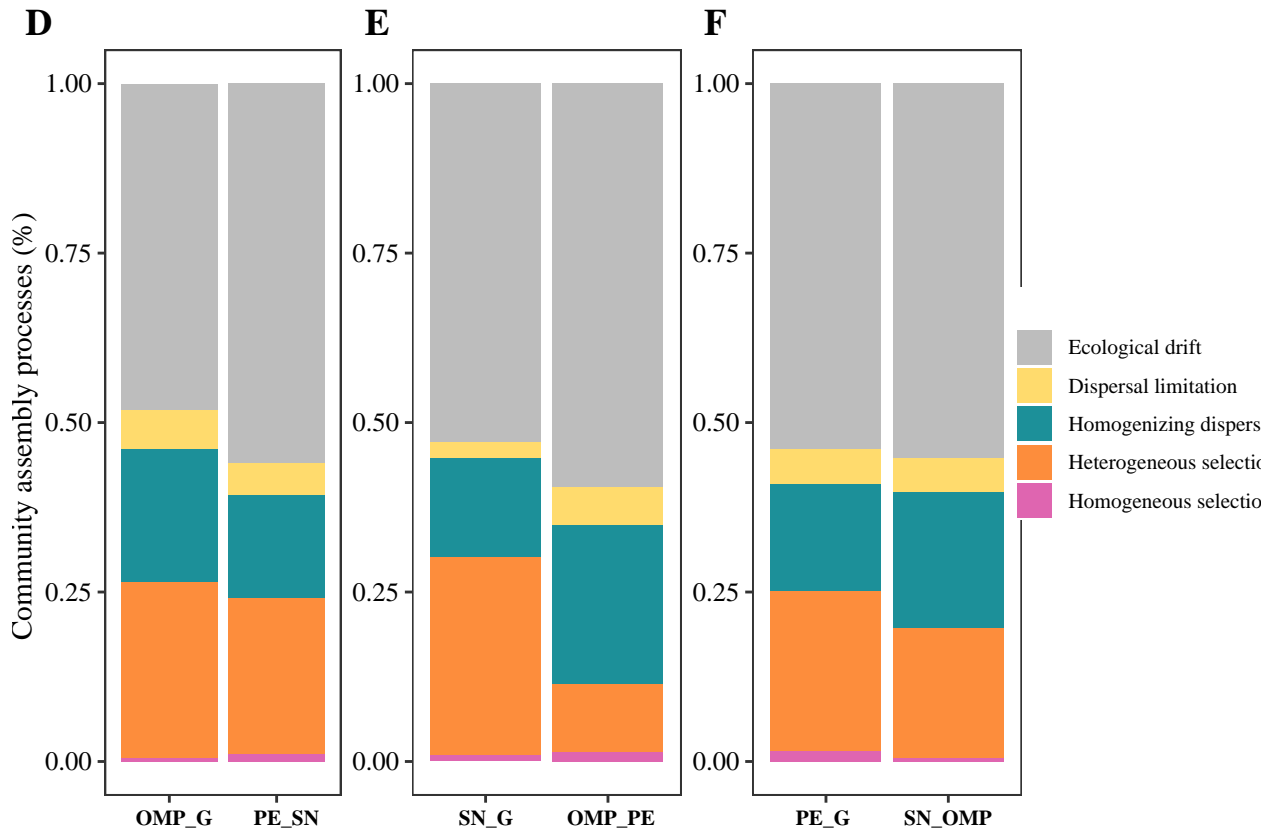
```

legend.position = "right", legend.box = "horizontal",
legend.key = element_rect(color = NA, fill = NA),
legend.spacing.x = unit(0.3, "cm"), legend.text=element_text(size=8)
legenda_prop.bRC_pp3=get_legend(prop.bRC_pp3)
prop.bRC_pp3=prop.bRC_pp3+theme(legend.position = "none")

```

Display Figure S2

Figure S2: A different approach, grouping multiple landscape estimates, shows the variability in the assembly processes of diatom communities at the regional scale. Representation of the various combinations of national parks (A, B, C). Percentage of community assembly processes in the six different combinations (D, E, F). The code represented the national parks' names: Ordesa y Monte Perdido (OMP), Guadarrama (G), Picos de Europa (PE), and Sierra Nevada (SN).



Percentages of community assembly processes

```
percentage=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="Percentage",na=""))
```

#Mean and standard deviation

```
tapply(percentage$perc.RC,interaction(percentage$category),mean)
```

```
##      Dispersal limitation      Ecological drift Heterogeneous selection
##      0.02129647          0.36171625          0.14656093
##      Homogeneous selection Homogenizing dispersal
##      0.02293006          0.44749629
```

```

tapply(percentage$perc.RC,interaction(percentage$category),sd)

##      Dispersal limitation      Ecological drift Heterogeneous selection
##      0.04122051              0.22029302      0.12664971
##      Homogeneous selection Homogenizing dispersal
##      0.03159312              0.28510311

tapply(percentage$perc.RC,interaction(percentage$proc),mean)

## Deterministic      Stochastic
##      0.1055024      0.2688316

tapply(percentage$perc.RC,interaction(percentage$proc),sd)

## Deterministic      Stochastic
##      0.1407632      0.2795182

tapply(percentage$perc.RC,interaction(percentage$group,percentage$proc),mean)

## Landscape.Deterministic      Local.Deterministic      Regional.Deterministic
##      0.20692894              0.06944444              0.12070174
##      Landscape.Stochastic      Local.Stochastic      Regional.Stochastic
##      0.19153130              0.28703704              0.25286551

tapply(percentage$perc.RC,interaction(percentage$group,percentage$proc),sd)

## Landscape.Deterministic      Local.Deterministic      Regional.Deterministic
##      0.2031950              0.1093632              0.1187036
##      Landscape.Stochastic      Local.Stochastic      Regional.Stochastic
##      0.1837350              0.3108396              0.2146375

tapply(percentage$perc.RC,interaction(percentage$group,percentage$category),mean)

##      Landscape.Dispersal limitation      Local.Dispersal limitation
##      0.01982735              0.01481481
##      Regional.Dispersal limitation      Landscape.Ecological drift
##      0.04172087              0.48156502
##      Local.Ecological drift      Regional.Ecological drift
##      0.27777778              0.53363247
##      Landscape.Heterogeneous selection      Local.Heterogeneous selection
##      0.17753411              0.11296296
##      Regional.Heterogeneous selection      Landscape.Homogeneous selection
##      0.22670604              0.02179755
##      Local.Homogeneous selection      Regional.Homogeneous selection
##      0.02592593              0.01469745
##      Landscape.Homogenizing dispersal      Local.Homogenizing dispersal
##      0.29927597              0.56851852
##      Regional.Homogenizing dispersal
##      0.18324318

tapply(percentage$perc.RC,interaction(percentage$group,percentage$category),sd)

##      Landscape.Dispersal limitation      Local.Dispersal limitation
##      0.021662327              0.048805633
##      Regional.Dispersal limitation      Landscape.Ecological drift
##      0.011892392              0.027467275
##      Local.Ecological drift      Regional.Ecological drift
##      0.235147013              0.035810619

```



```

## Landscape.Heterogeneous selection      Local.Heterogeneous selection
##                                0.092398056                                0.138214047
## Regional.Heterogeneous selection      Landscape.Homogeneous selection
##                                0.063282165                                0.013748501
##      Local.Homogeneous selection      Regional.Homogeneous selection
##                                0.038865539                                0.005098309
## Landscape.Homogenizing dispersal      Local.Homogenizing dispersal
##                                0.113788250                                0.284793750
## Regional.Homogenizing dispersal
##                                0.036059804

#Normality test and variance homogenization of variance
tapply(sqrt(percentage$perc.RC),interaction(percentage$group,percentage$proc),shapiro.test)

## $Landscape.Deterministic
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.87965, p-value = 0.103
##
##
## $Local.Deterministic
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.77991, p-value = 6.658e-06
##
##
## $Regional.Deterministic
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.80857, p-value = 0.01175
##
##
## $Landscape.Stochastic
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.91358, p-value = 0.3418
##
##
## $Local.Stochastic
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.84751, p-value = 6.778e-06
##
##
## $Regional.Stochastic

```

```

##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.87466, p-value = 0.02125
bartlett.test(sqrt(percentage$perc.RC)~interaction(percentage$group,percentage$proc))

##
##  Bartlett test of homogeneity of variances
##
## data:  sqrt(percentage$perc.RC) by interaction(percentage$group, percentage$proc)
## Bartlett's K-squared = 17.307, df = 5, p-value = 0.003953
#make subgroups to evaluate statistical differences between
#stochastic and deterministic processes
mire=percentage[1:90,]
n_park=percentage[91:110,]
p_iberica=percentage[11:140,]

#mire
tapply(mire$perc.RC,interaction(mire$proc),shapiro.test)

## $Deterministic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.66709, p-value = 8.936e-08
##
##
## $Stochastic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.8442, p-value = 5.49e-06
bartlett.test(mire$perc.RC,mire$proc)

##
##  Bartlett test of homogeneity of variances
##
## data:  mire$perc.RC and mire$proc
## Bartlett's K-squared = 34.996, df = 1, p-value = 3.304e-09
wilcox.test(mire$perc.RC~mire$proc)

##
##  Wilcoxon rank sum test with continuity correction
##
## data:  mire$perc.RC by mire$proc
## W = 627, p-value = 0.002942
## alternative hypothesis: true location shift is not equal to 0
#Landscape
tapply(n_park$perc.RC,interaction(n_park$proc),shapiro.test)

```

```

## $Deterministic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.82839, p-value = 0.02224
##
##
## $Stochastic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.87215, p-value = 0.1296
bartlett.test(n_park$perc.RC,n_park$proc)

##
##  Bartlett test of homogeneity of variances
##
## data:  n_park$perc.RC and n_park$proc
## Bartlett's K-squared = 0.084499, df = 1, p-value = 0.7713
t.test(n_park$perc.RC~n_park$proc)

##
##  Welch Two Sample t-test
##
## data:  n_park$perc.RC by n_park$proc
## t = 0.17774, df = 17.779, p-value = 0.8609
## alternative hypothesis: true difference in means between group Deterministic and group Stochastic is
## 95 percent confidence interval:
## -0.1667634  0.1975587
## sample estimates:
## mean in group Deterministic      mean in group Stochastic
##           0.2069289                0.1915313

#Regional
tapply(p_iberica$perc.RC,interaction(p_iberica$proc),shapiro.test)

## $Deterministic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.7768, p-value = 9.763e-08
##
##
## $Stochastic
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.87032, p-value = 1.601e-06

```

```

bartlett.test(p_iberica$perc.RC,p_iberica$proc)

##
## Bartlett test of homogeneity of variances
##
## data: p_iberica$perc.RC and p_iberica$proc
## Bartlett's K-squared = 21.839, df = 1, p-value = 2.965e-06

wilcox.test(p_iberica$perc.RC~p_iberica$proc)

##
## Wilcoxon rank sum test with continuity correction
##
## data: p_iberica$perc.RC by p_iberica$proc
## W = 1456.5, p-value = 0.003946
## alternative hypothesis: true location shift is not equal to 0
#inside scale evaluate the different processes
#mire
k<-kruskal.test(mire$perc.RC~mire$category)
k

##
## Kruskal-Wallis rank sum test
##
## data: mire$perc.RC by mire$category
## Kruskal-Wallis chi-squared = 51.199, df = 4, p-value = 2.028e-10

kruskalmc(mire$perc.RC~mire$category)

## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##


|                                                   | obs.dif   | critical.dif |
|---------------------------------------------------|-----------|--------------|
| ## Dispersal limitation-Ecological drift          | 31.666667 | 24.44431     |
| ## Dispersal limitation-Heterogeneous selection   | 18.416667 | 24.44431     |
| ## Dispersal limitation-Homogeneous selection     | 5.472222  | 24.44431     |
| ## Dispersal limitation-Homogenizing dispersal    | 52.083333 | 24.44431     |
| ## Ecological drift-Heterogeneous selection       | 13.250000 | 24.44431     |
| ## Ecological drift-Homogeneous selection         | 26.194444 | 24.44431     |
| ## Ecological drift-Homogenizing dispersal        | 20.416667 | 24.44431     |
| ## Heterogeneous selection-Homogeneous selection  | 12.944444 | 24.44431     |
| ## Heterogeneous selection-Homogenizing dispersal | 33.666667 | 24.44431     |
| ## Homogeneous selection-Homogenizing dispersal   | 46.611111 | 24.44431     |


##


|                                                   | stat.signif |
|---------------------------------------------------|-------------|
| ## Dispersal limitation-Ecological drift          | TRUE        |
| ## Dispersal limitation-Heterogeneous selection   | FALSE       |
| ## Dispersal limitation-Homogeneous selection     | FALSE       |
| ## Dispersal limitation-Homogenizing dispersal    | TRUE        |
| ## Ecological drift-Heterogeneous selection       | FALSE       |
| ## Ecological drift-Homogeneous selection         | TRUE        |
| ## Ecological drift-Homogenizing dispersal        | FALSE       |
| ## Heterogeneous selection-Homogeneous selection  | FALSE       |
| ## Heterogeneous selection-Homogenizing dispersal | TRUE        |
| ## Homogeneous selection-Homogenizing dispersal   | TRUE        |


```

#Landscape

```
k<-kruskal.test(n_park$perc.RC~n_park$category)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: n_park$perc.RC by n_park$category
## Kruskal-Wallis chi-squared = 16.371, df = 4, p-value = 0.002559
```

```
kruskalmc(n_park$perc.RC~n_park$category)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
```

	obs.dif	critical.dif	stat.signif
## Dispersal limitation-Ecological drift	13.75	11.74266	TRUE
## Dispersal limitation-Heterogeneous selection	6.25	11.74266	FALSE
## Dispersal limitation-Homogeneous selection	0.25	11.74266	FALSE
## Dispersal limitation-Homogenizing dispersal	9.75	11.74266	FALSE
## Ecological drift-Heterogeneous selection	7.50	11.74266	FALSE
## Ecological drift-Homogeneous selection	13.50	11.74266	TRUE
## Ecological drift-Homogenizing dispersal	4.00	11.74266	FALSE
## Heterogeneous selection-Homogeneous selection	6.00	11.74266	FALSE
## Heterogeneous selection-Homogenizing dispersal	3.50	11.74266	FALSE
## Homogeneous selection-Homogenizing dispersal	9.50	11.74266	FALSE

#Regional

```
k<-kruskal.test(p_iberica$perc.RC~p_iberica$category)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: p_iberica$perc.RC by p_iberica$category
## Kruskal-Wallis chi-squared = 70.665, df = 4, p-value = 1.643e-14
```

```
kruskalmc(p_iberica$perc.RC~p_iberica$category)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
```

	obs.dif	critical.dif	stat.signif
## Dispersal limitation-Ecological drift	60.653846	29.32869	
## Dispersal limitation-Heterogeneous selection	32.153846	29.32869	
## Dispersal limitation-Homogeneous selection	4.750000	29.32869	
## Dispersal limitation-Homogenizing dispersal	66.576923	29.32869	
## Ecological drift-Heterogeneous selection	28.500000	29.32869	
## Ecological drift-Homogeneous selection	55.903846	29.32869	
## Ecological drift-Homogenizing dispersal	5.923077	29.32869	
## Heterogeneous selection-Homogeneous selection	27.403846	29.32869	
## Heterogeneous selection-Homogenizing dispersal	34.423077	29.32869	
## Homogeneous selection-Homogenizing dispersal	61.826923	29.32869	
##			stat.signif
## Dispersal limitation-Ecological drift		TRUE	
## Dispersal limitation-Heterogeneous selection		TRUE	

```
## Dispersal limitation-Homogeneous selection FALSE
## Dispersal limitation-Homogenizing dispersal TRUE
## Ecological drift-Heterogeneous selection FALSE
## Ecological drift-Homogeneous selection TRUE
## Ecological drift-Homogenizing dispersal FALSE
## Heterogeneous selection-Homogeneous selection FALSE
## Heterogeneous selection-Homogenizing dispersal TRUE
## Homogeneous selection-Homogenizing dispersal TRUE

percentage$category<-factor(percentage$category,
                             c("Ecological drift","Dispersal limitation",
                               "Homogenizing dispersal",
                               "Heterogeneous selection",
                               "Homogeneous selection"))

percentage$group<-factor(percentage$group,
                         c("Local","Landscape","Regional"))
```

Display panel Figure 2B

```
cbp1 <- c("#969696", "#f5c32c", "#1c9099", "#fd8d3c", "#df65b0",
          "#56B4E9", "#009E73", "#293352", "#F0E442", "#ff0000")

perc_tur=ggplot(percentage,aes(x=group, y = perc.RC,colour=category)) +
  geom_boxplot(width=0.6,outlier.colour =NA)+
  geom_jitter(position=position_dodge(0.6))+
  scale_color_manual(values = c(cbp1[c(1:5)]),name="",
                     labels=c(expression("Ecological drift"),
                                expression("Dispersal limitation"),
                                expression("Homogenizing dispersal"),
                                expression("Heterogeneous selection"),
                                expression("Homogeneous selection")))+
  labs(title="",y="Percentage (%)",x="Assembly process")+ylim(0,1.1)+
  theme_bw()+
  theme(panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x = element_text(size=10, color="black",angle =0,hjust=0.5,vjust =1),
        axis.title.x = element_blank(),
        axis.title.y = element_blank(),axis.text.y =element_blank(),
        text = element_text(family="serif"))+
  theme(plot.margin = unit(c(-0.5,0.15, 0.1,0), "cm"))+
  guides(color=guide_legend(ncol = 2))+
  theme(legend.key.size = unit(0.8, "lines"),
        legend.margin=margin(0,0,0,0),legend.box.margin=margin(-10,-10,-10,-10),
        legend.position = "right",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.05,"cm"),legend.spacing.y = unit(-1, "cm"),
        legend.text=element_text(size=10))
legenda_perc_tur=get_plot_component(perc_tur, 'guide-box-right', return_all = TRUE)
perc_tur=perc_tur+theme(legend.position = "none")
```

Display panel Figure 2A

```

cbp1 <- c("#0072B2", "#e7298a", "#fc4e2a", "#bd0026", "#88419d", "#293352", "#238b45",
"#0072B2", "#E69F00", "#56B4E9", "#FFDB6D", "#009E73", "#CC79A7",
"#293352", "#F0E442", "#ff0000")

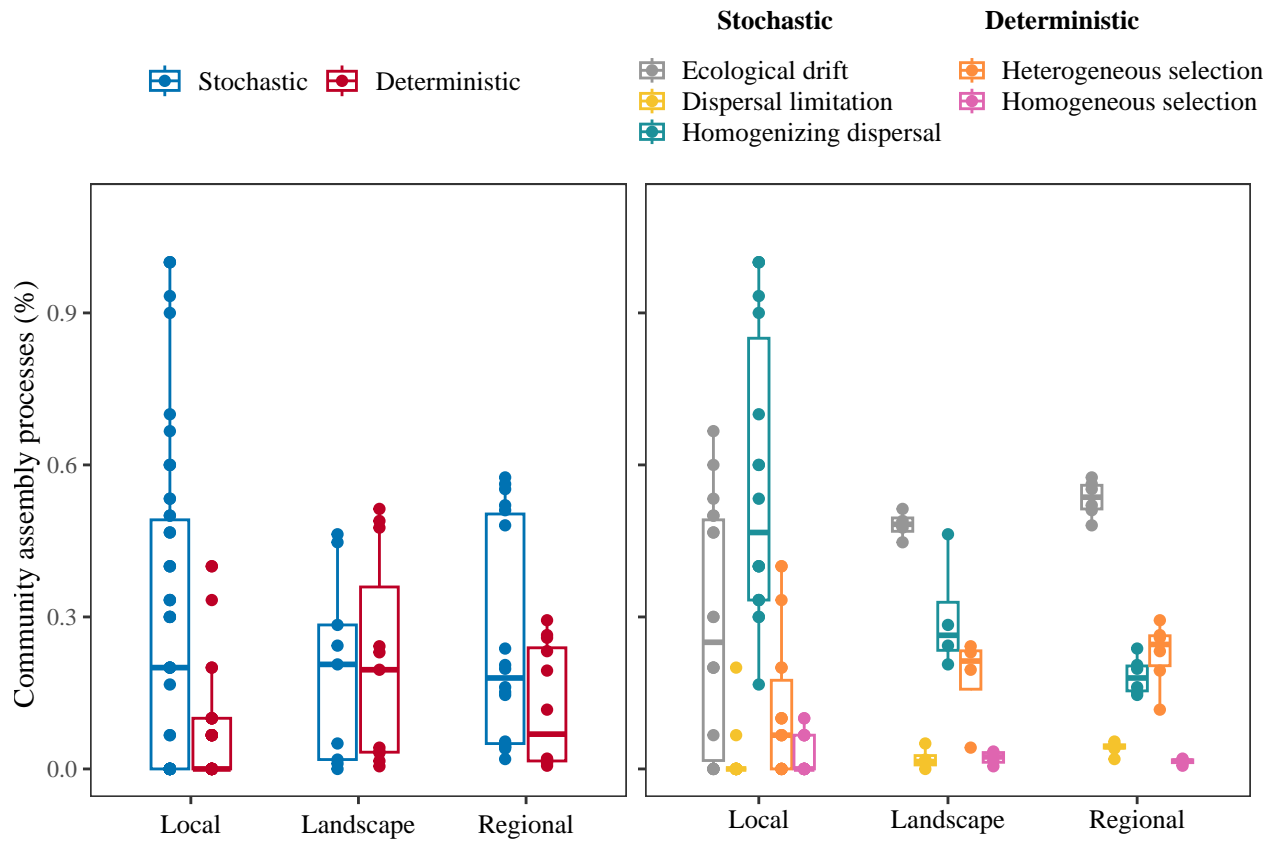
percentage$proc<-factor(ppercentage$proc,
                        c("Stochastic", "Deterministic"))

perc_proc=ggplot(ppercentage, aes(x=group, y = perc.RC, colour=proc)) +
  geom_boxplot(width=0.5, outlier.colour =NA)+
  geom_jitter(position=position_dodge(0.5))+
  scale_color_manual(values = c(cbp1[c(8,4)]), name="")+
  labs(title="", y="Community assembly processes (%)", x="Assembly process")+ylim(0,1.1)+
  theme_bw()+
  theme(panel.grid.minor = element_blank(), panel.grid.major = element_blank(),
        axis.text.x = element_text(size=10, color="black", angle =0, hjust=0.5, vjust =0),
        axis.title.x = element_blank(),
        text = element_text(family="serif"))+
  theme(plot.margin = unit(c(-0.5,0.15, 0.1,0), "cm"))+
  theme(legend.key.size = unit(1, "lines"),
        legend.margin=margin(0,1,0,1), legend.box.margin=margin(-10,-10,-10,-10),
        legend.position = "top", legend.box = "horizontal",
        legend.key = element_rect(color = NA, fill = NA),
        legend.spacing.x = unit(0.05, "cm"), legend.text=element_text(size=10))
legenda_perc_proc=get_plot_component(perc_proc, 'guide-box-top', return_all = TRUE)
perc_proc=perc_proc+theme(legend.position = "none")

```

Figure 2

Figure 2: Boxplots showing the percentage of variation in community organization accounted by different assembly mechanisms across the three spatial scales: local (i.e., individual mires), landscape (i.e., encompassing an entire national park), and regional (i.e., combining multiple national parks).



Analysis inside processes

```
#deterministic and stochastic
deter=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="deter_scale",na=""))
stoch=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="stoch_scale",na=""))
```

```
k<-kruskal.test(deter$perc.RC~deter$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: deter$perc.RC by deter$group
## Kruskal-Wallis chi-squared = 9.8158, df = 2, p-value = 0.007388
```

```
kruskalmc(deter$perc.RC~deter$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##
```

	obs.dif	critical.dif	stat.signif
## Landscape-Local	16.116162	14.16553	TRUE
## Landscape-Regional	4.393939	17.16360	FALSE
## Local-Regional	11.722222	13.70598	FALSE

```
k<-kruskal.test(stoch$perc.RC~stoch$group)
k
```



```
##
## Kruskal-Wallis rank sum test
##
## data:  stoch$perc.RC by stoch$group
## Kruskal-Wallis chi-squared = 0.58084, df = 2, p-value = 0.7479
```

```
kruskalmc(stoch$perc.RC~stoch$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##               obs.dif critical.dif stat.signif
## Landscape-Local    2.287037    20.27829      FALSE
## Landscape-Regional  6.388889    22.99343      FALSE
## Local-Regional     4.101852    15.32895      FALSE
```

Deterministic processes

```
#Heterogeneous selection and homogeneous selection
hete_selec=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="he.sele_scale",na=""))
homo_selec=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="ho.sele_scale",na=""))
```

```
k<-kruskal.test(hete_selec$perc.RC~hete_selec$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data:  hete_selec$perc.RC by hete_selec$group
## Kruskal-Wallis chi-squared = 5.5079, df = 2, p-value = 0.06367
```

```
kruskalmc(hete_selec$perc.RC~hete_selec$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##               obs.dif critical.dif stat.signif
## Landscape-Local    4.805556    10.885615      FALSE
## Landscape-Regional  3.916667    12.711660      FALSE
## Local-Regional     8.722222     9.283284      FALSE
```

```
k<-kruskal.test(homo_selec$perc.RC~homo_selec$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data:  homo_selec$perc.RC by homo_selec$group
## Kruskal-Wallis chi-squared = 2.3123, df = 2, p-value = 0.3147
```

```
kruskalmc(homo_selec$perc.RC~homo_selec$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##               obs.dif critical.dif stat.signif
## Landscape-Local    5.416667    10.885615      FALSE
```

```
## Landscape-Regional 1.250000 12.711660 FALSE
## Local-Regional 4.166667 9.283284 FALSE
```

Stochastic processes

```
#Homogenizing dispersal, dispersal limitation and ecological drift
Homo_disp=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="h.disp_scale",na=""))
Disp_lim=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="d.limi_scale",na=""))
Eco_drift=data.frame(read_xlsx("data\\Assembly.xlsx",sheet="e.dri_scale",na=""))
```

```
k<-kruskal.test(Homo_disp$perc.RC~Homo_disp$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: Homo_disp$perc.RC by Homo_disp$group
## Kruskal-Wallis chi-squared = 14.767, df = 2, p-value = 0.0006214
kruskalmc(Homo_disp$perc.RC~Homo_disp$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##
```

	obs.dif	critical.dif	stat.signif
## Landscape-Local	7.416667	10.885615	FALSE
## Landscape-Regional	7.083333	12.711660	FALSE
## Local-Regional	14.500000	9.283284	TRUE

```
k<-kruskal.test(Disp_lim$perc.RC~Disp_lim$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: Disp_lim$perc.RC by Disp_lim$group
## Kruskal-Wallis chi-squared = 12.482, df = 2, p-value = 0.001948
kruskalmc(Disp_lim$perc.RC~Disp_lim$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##
```

	obs.dif	critical.dif	stat.signif
## Landscape-Local	6.694444	10.885615	FALSE
## Landscape-Regional	4.916667	12.711660	FALSE
## Local-Regional	11.611111	9.283284	TRUE

```
k<-kruskal.test(Eco_drift$perc.RC~Eco_drift$group)
k
```

```
##
## Kruskal-Wallis rank sum test
##
## data: Eco_drift$perc.RC by Eco_drift$group
## Kruskal-Wallis chi-squared = 7.7564, df = 2, p-value = 0.02069
```

```
kruskalmc(Eco_drift$perc.RC~Eco_drift$group)
```

```
## Multiple comparison test after Kruskal-Wallis
## alpha: 0.05
## Comparisons
##               obs.dif critical.dif stat.signif
## Landscape-Local    4.694444    10.885615      FALSE
## Landscape-Regional  5.916667    12.711660      FALSE
## Local-Regional     10.611111     9.283284       TRUE
```

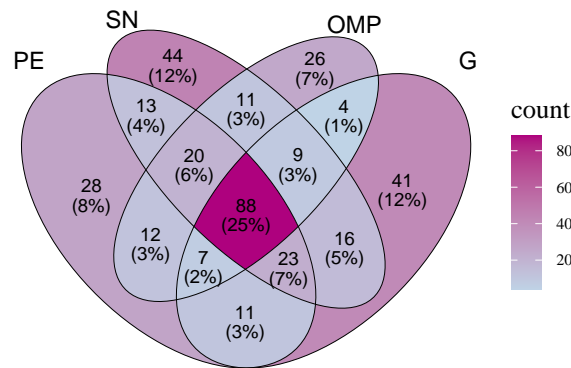
Venn diagram

```
#Load dataframe
Abun_relative=as.data.frame(read_xlsx("data\\Assembly.xlsx",
                                     sheet="venn_diag_dataframe",na=""))
rownames(Abun_relative)=Abun_relative$Species

Abun_relative[Abun_relative>0]=1
Abun_relative=Abun_relative[, -1]
```

Display panel Figure S4

Figure S4: 354 diatom species were identified, representing 60 genera across the four national parks (Table S2). 25% of these species had widespread distribution at the regional scale and comprised almost 70% of the total relative abundance (Table S3). Taxa uniqueness varied across landscapes; Ordesa y Monte Perdido exhibited the lowest percentage of unique species (7%, n=26), followed by Picos de Europa (8%, n=28), Guadarrama (12%, n=41), and the Sierra Nevada (13%, n=45). In addition to showing low occurrences and being relatively uncommon, these unique species were also characterized by their low relative abundances (Table S3).



Traits

```
common_sp=as.data.frame(read_xlsx("data\\Traits.xlsx",sheet="traits_common",na=""))
unique_sp=as.data.frame(read_xlsx("data\\Traits.xlsx",sheet="traits_unique",na=""))
```

```

row.names(common_sp)<-common_sp$Species
row.names(unique_sp)<-unique_sp$Species

#Reorder by abundance
unique_sp<-unique_sp[order(unique_sp$Mean,decreasing = TRUE),]
common_sp<-common_sp[order(common_sp$Mean,decreasing = TRUE),]

```

Display panel Figure S5

Figure S5: Diatom traits in A) common and B) unique species in the regional scale, and C) the measure of sizes. In black boxes, the species presents these traits. The heatmap traits are grouped by cluster.

Display panel Figure S5 A

```

common=heatmap(as.matrix(common_sp[,c(9:33)]),
               cluster_rows = FALSE,color=colorRampPalette(c("white","black"))(50),
               cellwidth = 5, cellheight = 5, fontsize = 7, legend = FALSE,
               fontfamily = "serif",
               fontface="italic")
common

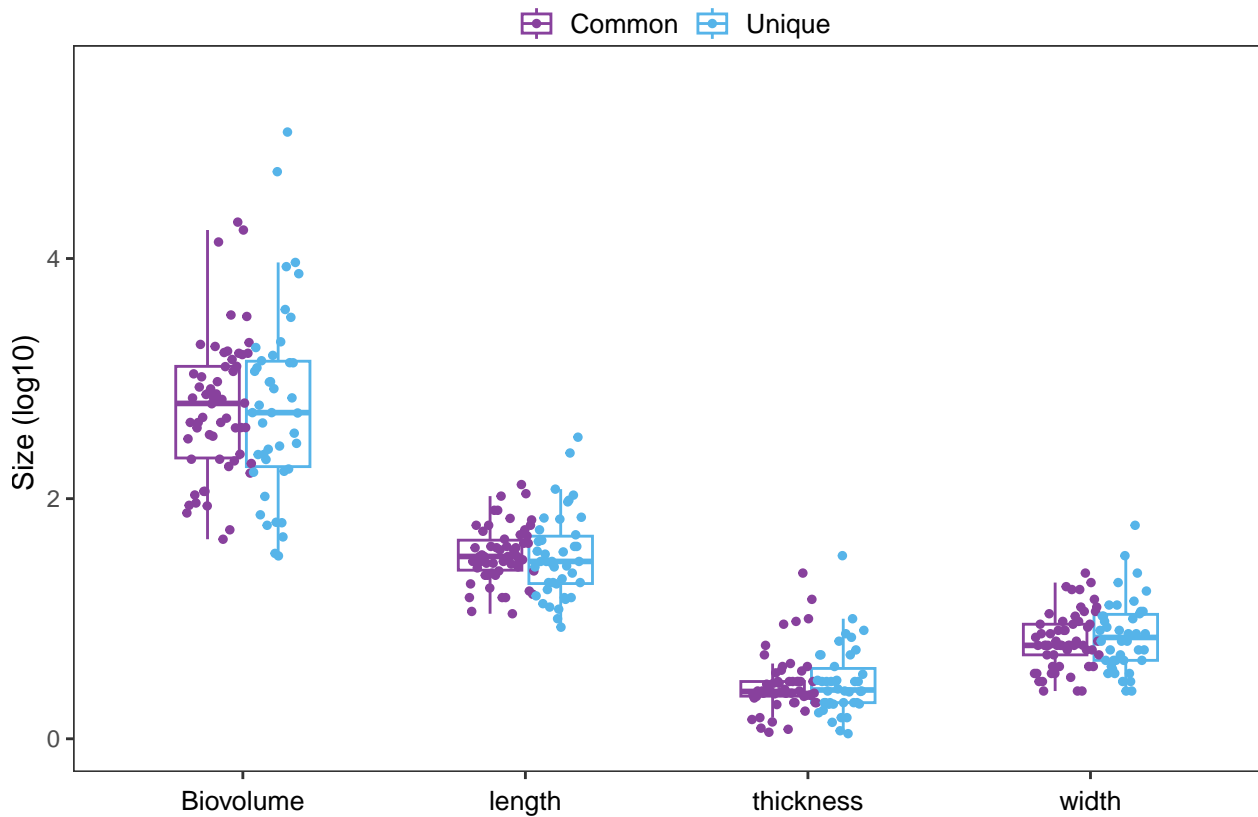
```



```

axis.title.x = element_blank()+
theme(plot.margin = unit(c(0,0, 0.5,0), "cm"))+
theme(legend.key.size = unit(1, "lines"),
      legend.margin=margin(0,1,0,1),legend.box.margin=margin(-10,-10,-10,-10),
      legend.position = "top",legend.box = "horizontal",
      legend.key = element_rect(color = NA, fill = NA),
      legend.spacing.x = unit(0.05,"cm"),legend.text=element_text(size=10))
sizes

```



Normality test

```

tapply(traits_size$Size,traits_size$Category,shapiro.test)

## $Biovolume
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.22504, p-value < 2.2e-16
##
##
## $length
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]

```

```
## W = 0.59722, p-value = 3.813e-15
##
##
## $thickness
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.46974, p-value < 2.2e-16
##
##
## $width
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.63585, p-value = 2.176e-14
bartlett.test(traits_size$Size~interaction(traits_size$State, traits_size$Category))

##
## Bartlett test of homogeneity of variances
##
## data:  traits_size$Size by interaction(traits_size$State, traits_size$Category)
## Bartlett's K-squared = 3740.9, df = 7, p-value < 2.2e-16
```

Statistical test

```
wilcox.test(traits_size$Size[traits_size$Category=="Biovolume"]~
            traits_size$State[traits_size$Category=="Biovolume"])

##
## Wilcoxon rank sum test with continuity correction
##
## data:  traits_size$Size[traits_size$Category == "Biovolume"] by traits_size$State[traits_size$Category == "Biovolume"]
## W = 1246.5, p-value = 0.845
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(traits_size$Size[traits_size$Category=="length"]~
            traits_size$State[traits_size$Category=="length"])

##
## Wilcoxon rank sum test with continuity correction
##
## data:  traits_size$Size[traits_size$Category == "length"] by traits_size$State[traits_size$Category == "length"]
## W = 1335, p-value = 0.4156
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(traits_size$Size[traits_size$Category=="thickness"]~
            traits_size$State[traits_size$Category=="thickness"])

##
## Wilcoxon rank sum test with continuity correction
##
## data:  traits_size$Size[traits_size$Category == "thickness"] by traits_size$State[traits_size$Category == "thickness"]
## W = 1174, p-value = 0.7609
```



```
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(traits_size$Size[traits_size$Category=="width"]~
            traits_size$State[traits_size$Category=="width"])

##
## Wilcoxon rank sum test with continuity correction
##
## data: traits_size$Size[traits_size$Category == "width"] by traits_size$State[traits_size$Category == "width"]
## W = 1146.5, p-value = 0.6195
## alternative hypothesis: true location shift is not equal to 0
```